

# Approximate pricing of European and Barrier claims in a local-stochastic volatility setting

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# Problem statement

We are interested in computing the price of a barrier-style claim

$$V = (V_t)_{0 \leq t \leq T} \quad (\text{option})$$

written on an asset

$$S = (S_t)_{0 \leq t \leq T} \quad (\text{asset})$$

whose payoff at the maturity date  $T$  is given by

$$\mathbb{1}_{\{\tau > T\}} \hat{\varphi}(S_T), \quad \tau = \inf\{t \geq 0 : S_t \notin \hat{I}\}. \quad (\text{payoff})$$

where  $\hat{I}$  is an interval in  $\mathbb{R}$ .

- The option becomes worthless if  $S$  leaves  $\hat{I}$  at any time  $t \leq T$ .
- These types of options are known as *knock-out* options.

# Problem statement

Examples:

- $\hat{I} = (L, U)$  - double-barrier knock-out
- $\hat{I} = (L, \infty)$  - single-barrier option with lower barrier
- $\hat{I} = (-\infty, U)$  - single-barrier option with upper barrier
- $\hat{I} = (-\infty, \infty)$  - European option

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We can price knock-in options by pricing European and knock-out options using knock-in knock-out parity

$$V_{\hat{I}}^{(\text{knock-in})} + V_{\hat{I}}^{(\text{knock-out})} = V^{(\text{European})},$$

where the payoff of a knock-in option is given by

$$\mathbb{1}_{\{\tau \leq T\}} \hat{\varphi}(S_T).$$

# Asset model

For an asset  $S$ , we consider models of in a general **local-stochastic** volatility setting

$$\begin{aligned}S_t &= e^{X_t}, \\dX_t &= \mu(X_t, Y_t)dt + \sigma(X_t, Y_t)dW_t, \\dY_t &= c(X_t, Y_t)dt + g(X_t, Y_t)dB_t, \\d\langle W, B \rangle_t &= \rho dt,\end{aligned}$$

where  $W$  and  $B$  are correlated Brownian motions under the pricing probability measure  $\mathbb{P}$ .

# Risk-neutral price

Let

- $r = 0$ ,
- $I = \log \hat{I}$ ,
- $\varphi(x) = \hat{\varphi}(e^x) = \hat{\varphi}(s)$ .

To avoid arbitrage, all traded assets must be martingales under the pricing measure  $\mathbb{P}$ . The value  $V_t$  of the claim with the payoff

$$\mathbb{1}_{\{\tau > T\}} \varphi(X_T), \quad \tau = \inf\{t \geq 0 : X_t \notin I\} \quad (\text{payoff})$$

at time  $t \leq T$  is given by

$$V_t = \mathbb{1}_{\{\tau > t\}} u(t, X_t, Y_t),$$

where

$$u(t, x, y) := \mathbb{E} \left( \mathbb{1}_{\{\tau > T\}} \varphi(X_T) \mid X_t = x, Y_t = y, \tau > t \right).$$

# Possible Approaches

How might one solve the pricing problem?

- Simulation
  - Ex: Monte Carlo
  - **Limitation**: Simulation gives you the price for one  $(X_0, Y_0)$  and parameter choice.
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- Numerical PDE solver
  - Ex: Solve PDE using finite difference or finite element
  - **Limitation**: Numerical solvers suffer from the “curse of dimensionality.”
  - **Limitation**: Discretized solution



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  - **Limitation**: Discretized solution
- Analytical techniques on the PDE
  - Ex: perturbation theory
  - **Advantage**: Fast evaluation at higher dimension
  - **Advantage**: Ease of implementation

# Pricing PDE

The function  $u$

$$u(t, x, y) = \mathbb{E} \left( \mathbb{1}_{\{\tau > T\}} \varphi(X_T) | X_t = x, Y_t = y, \tau > t \right),$$

is the unique classical solution of the Kolmogorov Backward equation

$$0 = (\partial_t + \mathcal{A})u, \quad u(T, \cdot) = \varphi,$$

where  $\mathcal{A}$ , the **generator** of  $(X, Y)$ , is given explicitly by

$$\begin{aligned} \mathcal{A} = & \frac{1}{2} \sigma^2(x, y) \partial_x^2 + \rho \sigma(x, y) g(x, y) \partial_x \partial_y + \frac{1}{2} g^2(x, y) \partial_y^2 \\ & + \mu(x, y) \partial_x + c(x, y) \partial_y, \end{aligned}$$

and the domain of  $\mathcal{A}$  is given by

$$\text{dom}(\mathcal{A}) := \{g \in C^2 : \lim_{x \rightarrow \partial I} g(x, y) = 0\}.$$

# Our approach

The full pricing PDE

$$\begin{aligned} 0 = & \partial_t u + \frac{1}{2} \sigma^2(x, y) \partial_x^2 u + \rho \sigma(x, y) g(x, y) \partial_x \partial_y u \\ & + \frac{1}{2} g^2(x, y) \partial_y^2 u + \mu(x, y) \partial_x u + c(x, y) \partial_y u, \\ u(T, \cdot, \cdot) = & \varphi, \end{aligned}$$

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is not generally solvable in closed form.

If  $\sigma, g, \mu, c$  were constant and  $\rho = 0$ , the pricing PDE would be

$$\partial_t u + \frac{1}{2} \sigma^2 \partial_x^2 u + \frac{1}{2} g^2 \partial_y^2 u + \mu \partial_x u + c \partial_y u = 0,$$

which is solvable. This suggests a perturbation expansion...

# Perturbation framework

Let  $f \in \{\frac{1}{2}\sigma^2, \sigma g, \frac{1}{2}g^2, \mu, c\}$  and  $(\bar{x}, \bar{y}) \in I \times \mathbb{R}$ . We introduce  $\varepsilon \in [0, 1]$  and define

$$f^\varepsilon(x, y) := f(\bar{x} + \varepsilon(x - \bar{x}), \bar{y} + \varepsilon(y - \bar{y})).$$

Note that  $f^\varepsilon(x, y)|_{\varepsilon=1} = f(x, y)$  and  $f^\varepsilon(x, y)|_{\varepsilon=0} = f(\bar{x}, \bar{y})$ .

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Taylor expanding  $f^\varepsilon$  about the point  $\varepsilon = 0$  yields

$$f^\varepsilon = f_0 + \varepsilon f_1 + \varepsilon^2 f_2 + \cdots,$$

where

$$f_n(x, y) = \sum_{i=0}^n \frac{\partial_x^{n-i} \partial_y^i f(\bar{x}, \bar{y})}{i!(n-i)!} (x - \bar{x})^{n-i} (y - \bar{y})^i.$$

# Perturbation framework

Recall

$$\mathcal{A} = \frac{1}{2}\sigma^2(x, y)\partial_x^2 + \textcolor{red}{\rho}\sigma(x, y)g(x, y)\partial_x\partial_y + \frac{1}{2}g^2(x, y)\partial_y^2 \\ + \mu(x, y)\partial_x + c(x, y)\partial_y,$$

Replacing  $f \in \{\frac{1}{2}\sigma^2, \sigma g, \frac{1}{2}g^2, \mu, c\}$  with  $f^\epsilon$  in  $\mathcal{A}$  and expanding yields

$$\mathcal{A}^{\epsilon, \textcolor{red}{\rho}} = \sum_{n=0}^{\infty} \epsilon^n (\mathcal{A}_{n,0} + \textcolor{red}{\rho}\mathcal{A}_{n,1}),$$

where

$$\mathcal{A}_{n,0} := \left(\frac{1}{2}\sigma^2\right)_n \partial_x^2 + \left(\frac{1}{2}g^2\right)_n \partial_y^2 + \mu_n \partial_x + c_n \partial_y \\ \mathcal{A}_{n,1} := (\sigma g)_n \partial_x \partial_y.$$

# Perturbation framework

We try to solve

$$(\partial_t + \mathcal{A}^{\varepsilon, \rho}) u^{\varepsilon, \rho} = 0, \quad u^{\varepsilon, \rho}(T, \cdot, \cdot) = \varphi$$

by expanding  $u^{\varepsilon, \rho}$  in powers of  $\varepsilon$  and  $\rho$  as follows

$$u^{\varepsilon, \rho} = \sum_{n=0}^{\infty} \sum_{i=0}^n \varepsilon^{n-i} \rho^i u_{n-i, i}.$$

An approximation to the solution of the original pricing PDE

$$(\partial_t + \mathcal{A}) u = 0, \quad u(T, \cdot, \cdot) = \varphi$$

will be obtained by setting  $\varepsilon = 1$  in  $u^{\varepsilon, \rho}$ .



# Perturbation framework

We now have the parameterized set of PDEs

$$(\partial_t + \mathcal{A}^{\varepsilon, \rho}) u^{\varepsilon, \rho} = 0, \quad u^{\varepsilon, \rho}(T, \cdot, \cdot) = \varphi.$$

Inserting  $\mathcal{A}^{\varepsilon, \rho}$  and  $u^{\varepsilon, \rho}$  and collecting powers of  $\varepsilon$  and  $\rho$  gives

$$\begin{aligned} O(\varepsilon^0 \rho^0) : \quad & (\partial_t + \mathcal{A}_{0,0}) u_{0,0} = 0, \quad u_{0,0}(T, \cdot, \cdot) = \varphi, \\ O(\varepsilon^n \rho^k) : \quad & (\partial_t + \mathcal{A}_{0,0}) u_{n,k} + F_{n,k} = 0, \quad u_{n,k}(T, \cdot, \cdot) = 0, \end{aligned}$$

where

$$F_{n,k} = \sum_{i=0}^n \sum_{j=0}^k (1 - \delta_{i+j,0}) \mathcal{A}_{i,j} u_{n-i,k-j}.$$

- $O(\varepsilon^0 \rho^0)$  is a constant coefficient heat equation.
- $O(\varepsilon^n \rho^k)$  is a constant coefficient heat equation with a forcing term.

# $N$ th order approximation

## Definition

Let  $u$  be the unique classical solution of PDE problem (1).

$$(\partial_t + \mathcal{A}) u = 0, \quad u(T, \cdot, \cdot) = \varphi, \quad (1)$$

We define  $\bar{u}_N^{\rho}$ , the  $N$ th order approximation of  $u$ , as

$$\bar{u}_N^{\rho}(t, x, y) := \sum_{i=0}^N \sum_{j=0}^i \varepsilon^j \rho^{i-j} u_{j,i-j}(t, x, y) \Big|_{(\bar{x}, \bar{y}, \varepsilon) = (x, y, 1)},$$

where  $u_{0,0}$  satisfies (2) and  $u_{n,k}$  satisfies (3) for  $(n, k) \neq (0, 0)$ .

$$(\partial_t + \mathcal{A}_{0,0}) u_{0,0} = 0, \quad u_{0,0}(T, \cdot, \cdot) = \varphi, \quad (2)$$

$$(\partial_t + \mathcal{A}_{0,0}) u_{n,k} + F_{n,k} = 0, \quad u_{n,k}(T, \cdot, \cdot) = 0. \quad (3)$$

# Duhamel's principal

*Duhamel's principle* states that the the unique classical solution to

$$(\partial_t + \mathcal{A}_{0,0})u + F = 0, \quad u(T, \cdot, \cdot) = h,$$

is given by

$$u(t, x, y) = \mathcal{P}_{0,0}(t, T)h(x, y) + \int_t^T ds \mathcal{P}_{0,0}(t, s)F(s, x, y),$$

where we have introduced  $\mathcal{P}_{0,0}$  the *semigroup* generated by  $\mathcal{A}_{0,0}$ , which is defined as follows

$$\mathcal{P}_{0,0}(t, s)h(x, y) = \int_I d\xi \int_{\mathbb{R}} d\eta \mathbf{\Gamma}_{0,0}(t, x, y; s, \xi, \eta)h(\xi, \eta),$$

where  $0 \leq t \leq s \leq T$ , and  $\mathbf{\Gamma}_{0,0}$  is the solution of

$$0 = (\partial_t + \mathcal{A}_{0,0})\mathbf{\Gamma}_{0,0}(\cdot, \cdot, \cdot; T, \xi, \eta), \quad \mathbf{\Gamma}_{0,0}(T, \cdot, \cdot; T, \xi, \eta) = \delta_{\xi, \eta}.$$

# Formula for $u_{n,k}$

## Proposition

The function  $u_{0,0}$  is given by

$$u_{0,0}(t) = \mathcal{P}_{0,0}(t, T)\varphi,$$

and for  $(n, k) \neq (0, 0)$ , we have

$$u_{n,k}(t) = \sum_{j=1}^{n+k} \sum_{I_{n,k,j}} \int_t^T ds_1 \int_{s_1}^T ds_2 \cdots \int_{s_{j-1}}^T ds_j \\ \mathcal{P}_{0,0}(t, s_1) \mathcal{A}_{n_1, k_1} \cdots \mathcal{P}_{0,0}(s_{j-1}, s_j) \mathcal{A}_{n_j, k_j} \mathcal{P}_{0,0}(s_j, T) \varphi,$$

with  $I_{n,k,j}$  given by

$$I_{n,k,j} = \left\{ \begin{pmatrix} n_1, \dots, n_j \\ k_1, \dots, k_j \end{pmatrix} \in \mathbb{Z}_+^{2 \times j} \left| \begin{array}{l} n_1 + \cdots + n_j = n, \\ k_1 + \cdots + k_j = k, \\ 1 \leq n_i + k_i, \text{ for all } 1 \leq i \leq j \end{array} \right. \right\}.$$

# Asymptotic accuracy for European claims

Let  $I = \mathbb{R}$  (European option), and let  $h - 1$  be the number of Lipschitz continuous derivatives of  $\varphi$ . Then under certain regularity assumptions on the coefficients  $(\mu, \sigma, g, c)$ , the approximate solution satisfies the following:

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$$|(u - \bar{u}_0^{\rho})(t, x, y)| \leq C (T - t)^{\frac{h+1}{2}}, \quad 0 \leq t < T, \quad x \in I, y \in \mathbb{R}.$$

For  $N \geq 1$ , we have

$$|(u - \bar{u}_N^{\rho})(t, x, y)| \leq C ((T - t)^{\frac{1}{2}} + |\rho|) \sum_{i=0}^N |\rho|^i (T - t)^{\frac{N-i+h}{2}}$$
$$0 \leq t < T, x \in I, y \in \mathbb{R}.$$

The positive constants  $C$  in depend only on  $N$ ,  $\varphi$  (and  $\sigma, g, \mu, c$ ).

## Numerical example: CEV model

Suppose that  $S = e^X$  has Constant Elasticity of Variance (Cox (1975)) dynamics i.e.

$$dS_t = \sigma S_t^\gamma dW_t,$$

$$dX_t = -\frac{1}{2}\sigma^2 e^{2X_t(\gamma-1)} dt + \sigma e^{X_t(\gamma-1)} dW_t.$$

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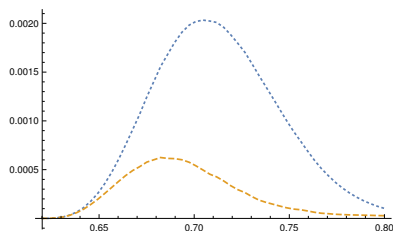
$$dX_t = -\frac{1}{2}\sigma^2 e^{2X_t(\gamma-1)} dt + \sigma e^{X_t(\gamma-1)} dW_t.$$

We consider double-barrier knock-out calls and puts with the following parameters fixed

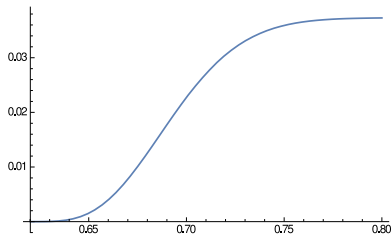
$X_0$	$K$	$T$	$\sigma$	$\gamma$
0.62	0.62	0.083	0.32	0.019



# CEV double-barrier call

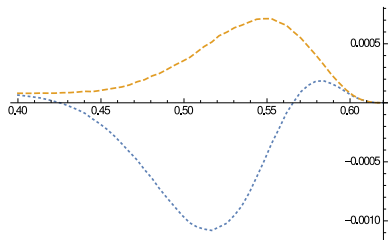


**Figure 1:** For the CEV with  $L = 0$ , we plot  $u - \bar{u}_0$  (blue dotted) and  $u - \bar{u}_2$  (orange dashed) as a function of the upper barrier  $U$  for a call option.

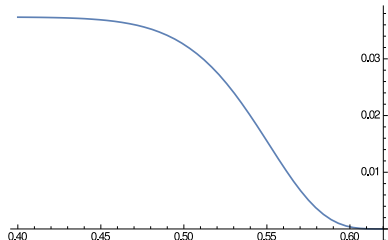


**Figure 2:** For the CEV model with  $L = 0$ , we plot  $u$  as a function of the upper barrier  $U$  for a call option.

# CEV double-barrier put



**Figure 3:** For the CEV model with  $U = 1$ , we plot  $u - \bar{u}_0$  (blue dotted) and  $u - \bar{u}_2$  (orange dashed) as a function of the lower barrier  $L$  for a put option.



**Figure 4:** For the CEV model with  $U = 1$ , we plot  $u$  as a function of the lower barrier  $L$  for a put option.

## Numerical example: Heston model

Suppose that  $S = e^X$  has Heston (Heston (1993)) dynamics i.e.

$$dS_t = \sqrt{Y_t} S_t dW_t,$$

$$dX_t = -\frac{1}{2} Y_t dt + \sqrt{Y_t} dW_t,$$

$$dY_t = \kappa(\theta - Y_t) dt + \delta \sqrt{Y_t} dB_t,$$

$$d\langle W, B \rangle_t = \rho dt$$

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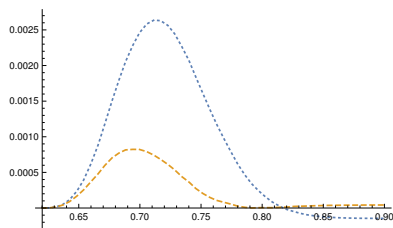
$$dY_t = \kappa(\theta - Y_t) dt + \delta \sqrt{Y_t} dB_t,$$

$$d\langle W, B \rangle_t = \rho dt$$

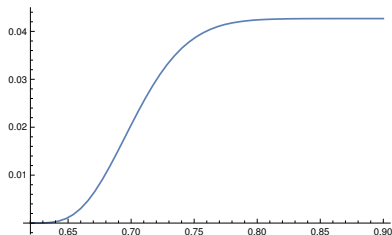
We specify a model

$X_0$	$Y_0$	$K$	$T$	$\rho$	$\kappa$	$\theta$	$\delta$
0.62	0.04	.62	0.083	-0.4	1.15	0.04	0.2

# Heston double-barrier call

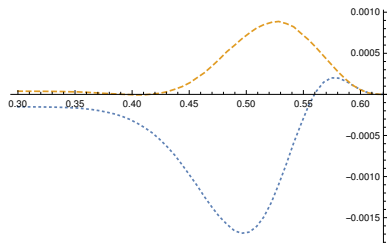


**Figure 5:** For the Heston model, we plot  $u - \bar{u}_0^\rho$  (blue dotted) and  $u - \bar{u}_2^\rho$  (orange dotted-dashed) as a function of the upper barrier  $U$  for a call option.

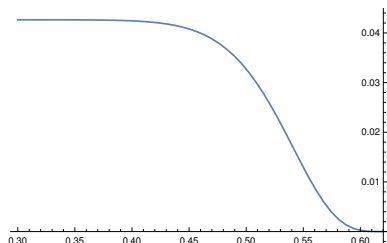


**Figure 6:** For the Heston model, we plot  $u$  as a function of the upper barrier  $U$  for a call option.

# Heston double-barrier put



**Figure 7:** For the Heston model, we plot  $u - \bar{u}_0^\rho$  (blue dotted) and  $u - \bar{u}_2^\rho$  (orange dotted-dashed) as a function of the lower barrier  $L$  for a put option.



**Figure 8:** For the Heston model, we plot  $u$  as a function of the lower barrier  $L$  for a put option.

# Conclusion

- Limitations of numerical methods and simulations
- Pricing options exactly under general dynamics is impossible, so we turn to asymptotics
- Constant coefficient PDE theory is used to solve the asymptotic problem
- Rigorous accuracy results for European options
- Numerical accuracy demonstrations for barrier options

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